

# Reconstruction of Evolving, Non-Convex Curves from a Sequence of Single-Angle Projections

-- or --

Radiographic Reconstructions Utilizing  
the Dynamics Connecting Differing Times

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Thomas J. Asaki and Kevin R. Vixie

Los Alamos National Laboratory, Los Alamos, NM

Erik M. Bollt

Clarkson University, Potsdam, NY

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## The Exploratory Concepts

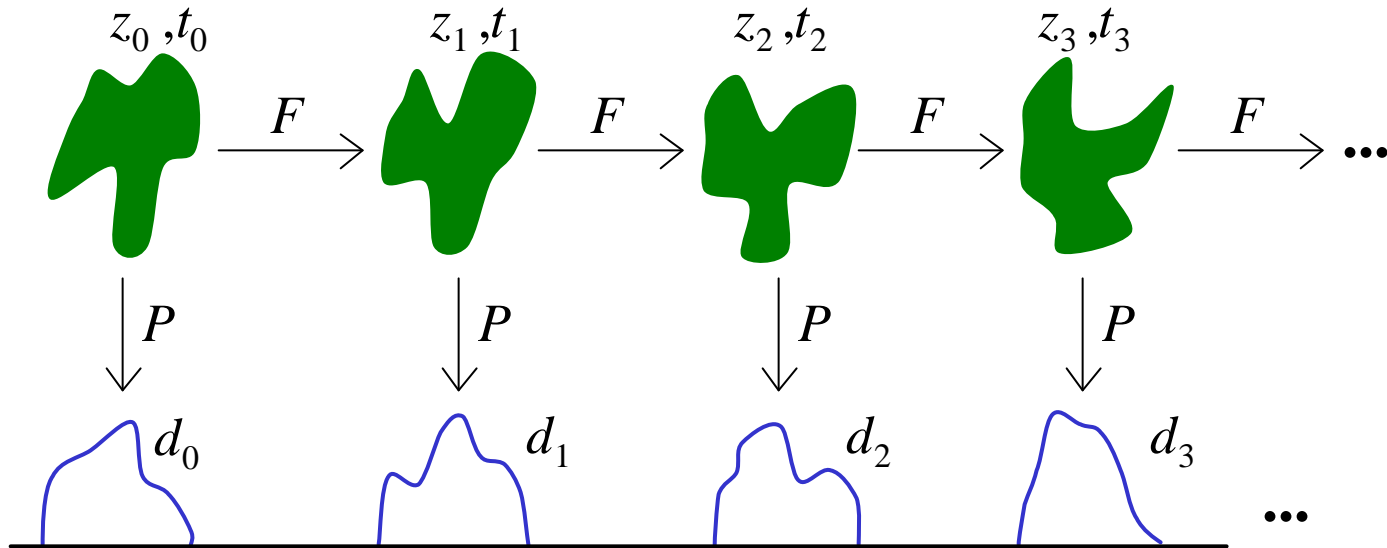
Can a time-sequence of single-view (or few) radiographs be used to enhance object reconstructions through dynamics constraints?

How much knowledge is required of the dynamics that connect different times?

Can the nature of the dynamics be determined simultaneously along with the object description?

How are these results affected by computational time, error propagation, and metrics?

# Incorporating Dynamics in Reconstructions



$z_0$  is some initial object

$z_k = F^k z_0$  is the evolved object at time  $t_k$

$d_k = P z_k$  is a low-dimensional projection of object  $z_k$

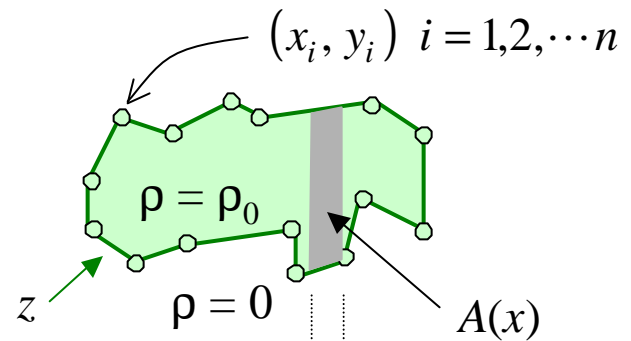
$F = F(\lambda, t)$  is a time evolution operator

$f$  = merit function 
$$f = \sum_k \|d_k - P F^k z_0\|$$

# A First Look

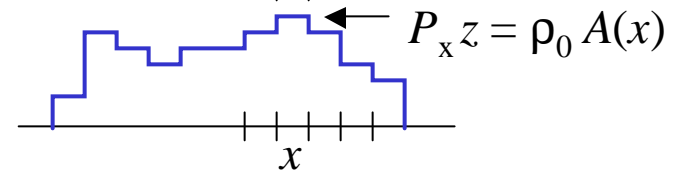
## The Object:

A uniform-density simply-connected object defined by a set of ordered points



## The Projection Data:

Single-view discrete mass projection, possibly with simulated noise

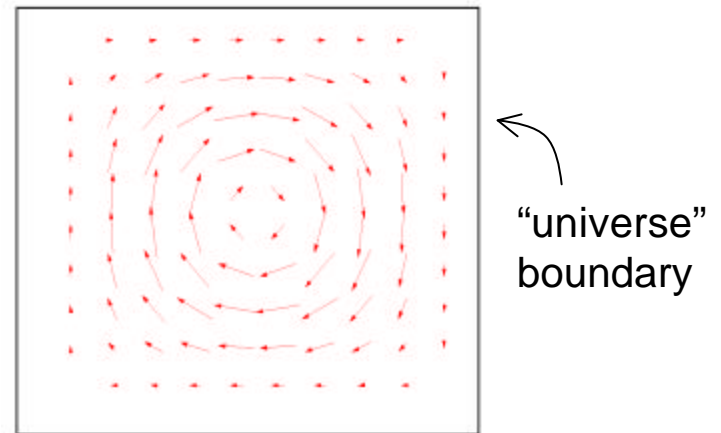


## The Dynamics:

A vortex-like fixed advection flow

$$\dot{x} = -\sin^2(x)\sin(2y)$$

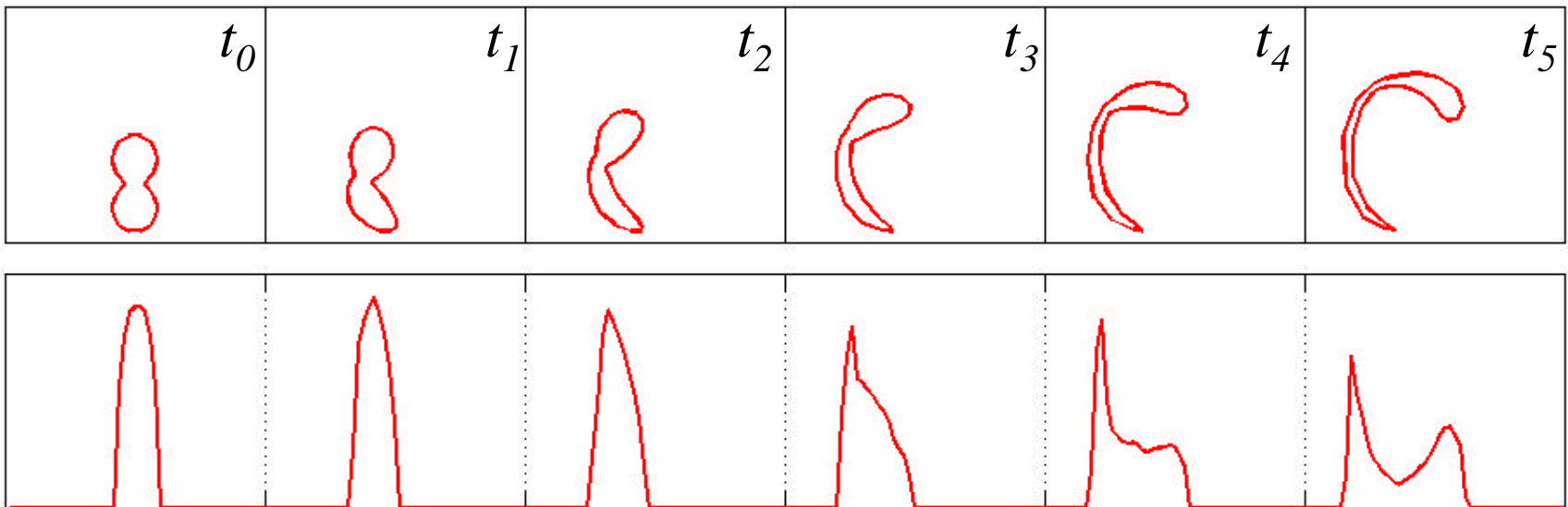
$$\dot{y} = +\sin^2(y)\sin(2x)$$



## Example Experiment and Data

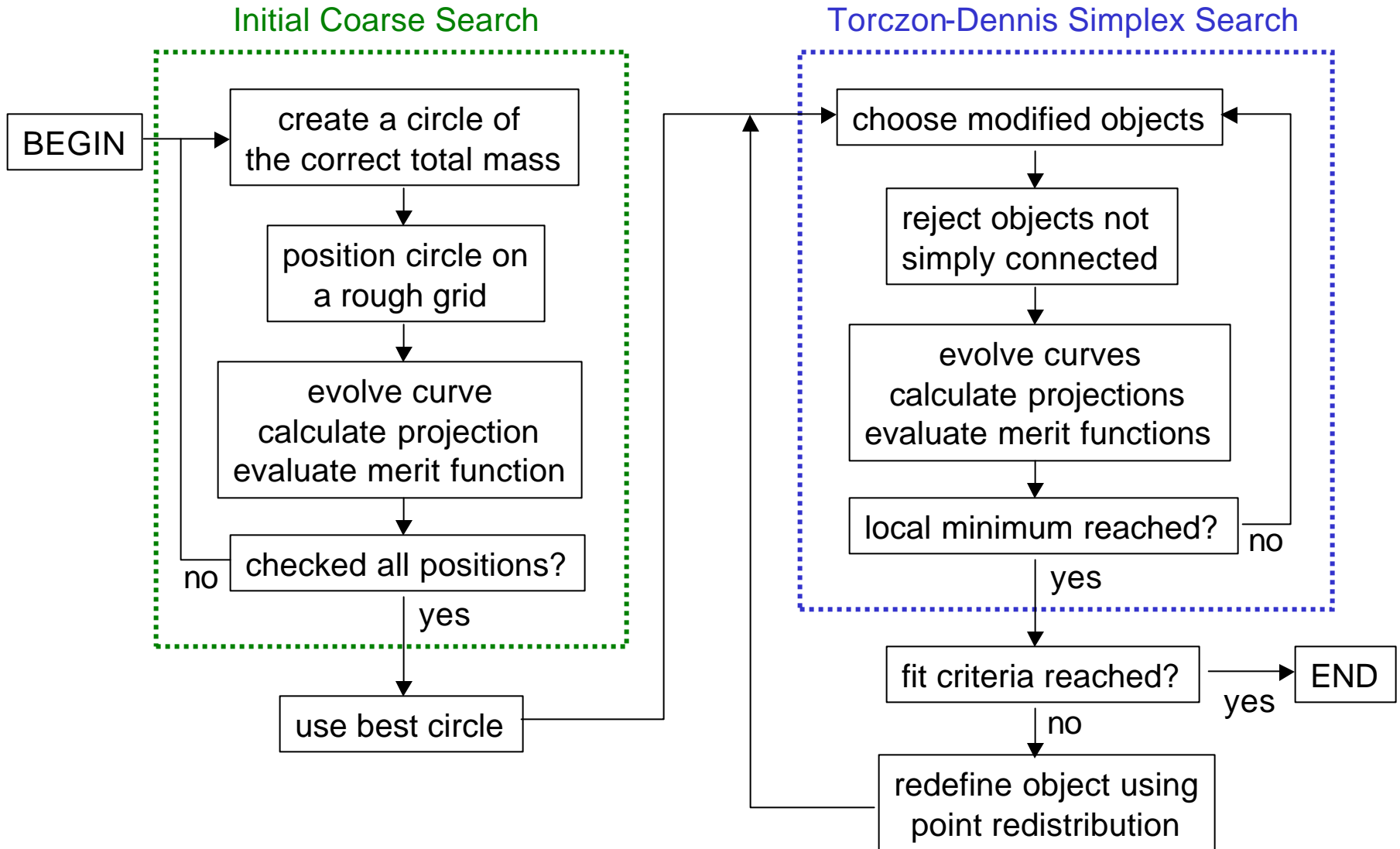
The object is defined by 30 vertices.  
Each of six time views has 50 sampling bins.  
The dynamics is exactly known.

Evolving object at six different times

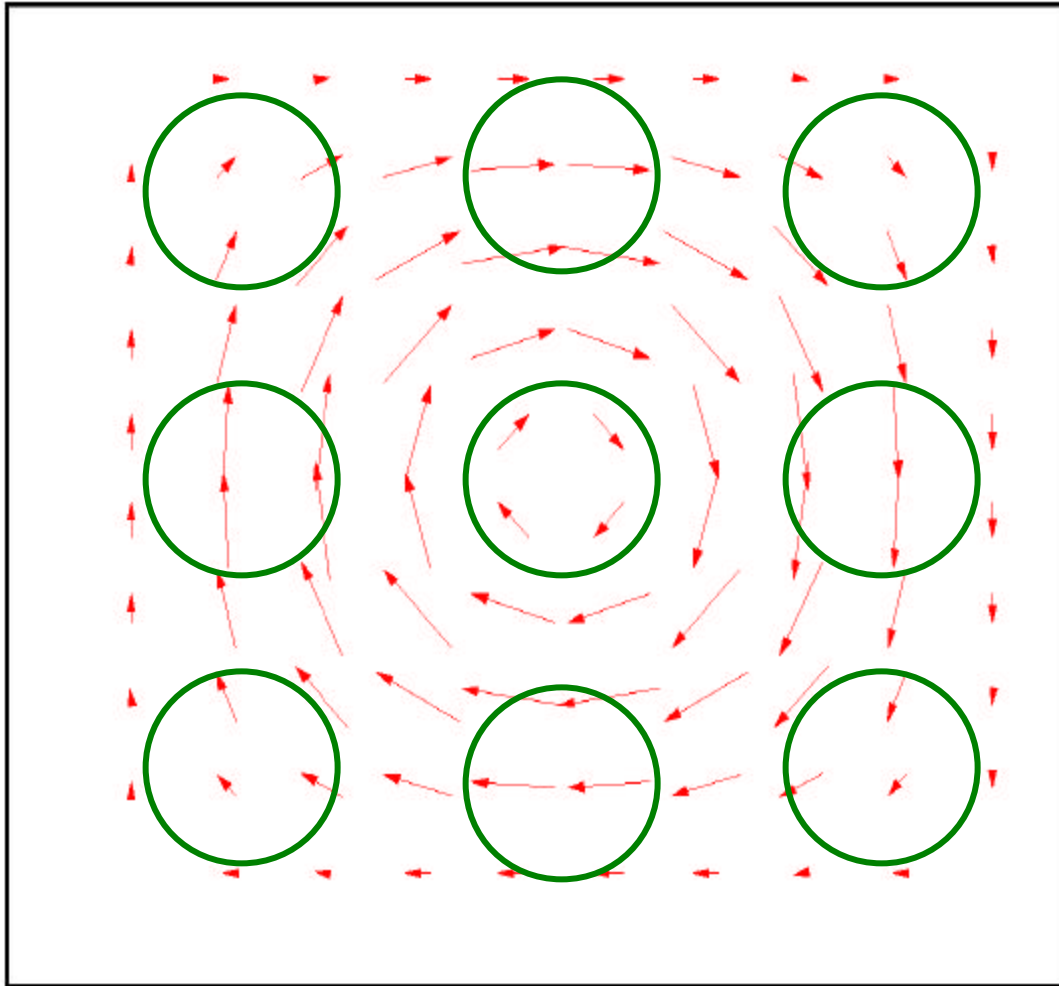


Corresponding projection data (noiseless case)

# The Algorithm



# Initial Coarse Search

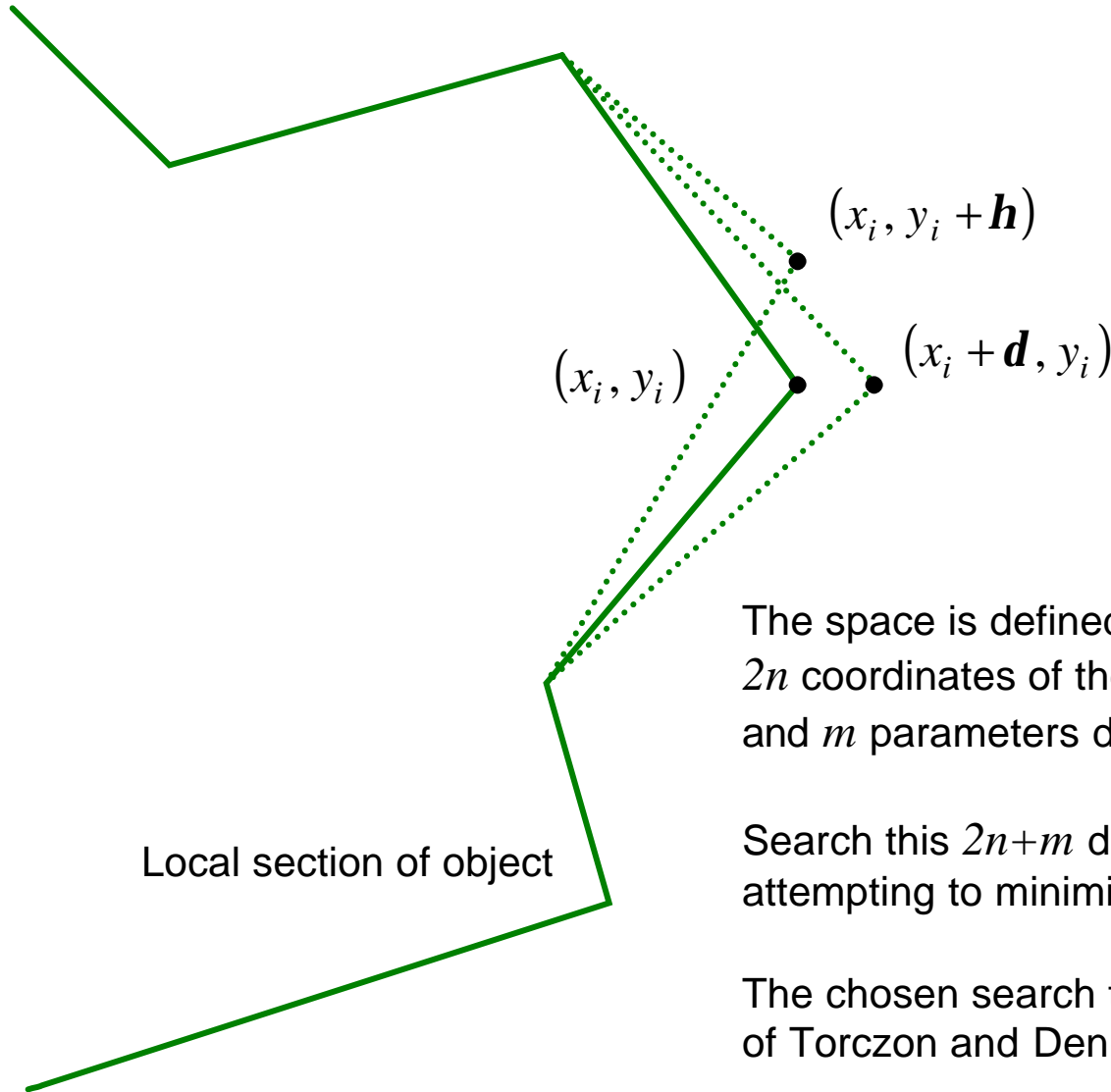


Begin with an initial circular object of the correct mass (regular polygon of  $n$  sides).

Search over small set of initial locations in the universe -- computing projections and evaluating the merit functions.

This quick and simple process determines the most likely initial location for the unknown object.

# The Refined Search



The space is defined by:  
 $2n$  coordinates of the vertices of the initial object  
and  $m$  parameters defining the evolution dynamics.

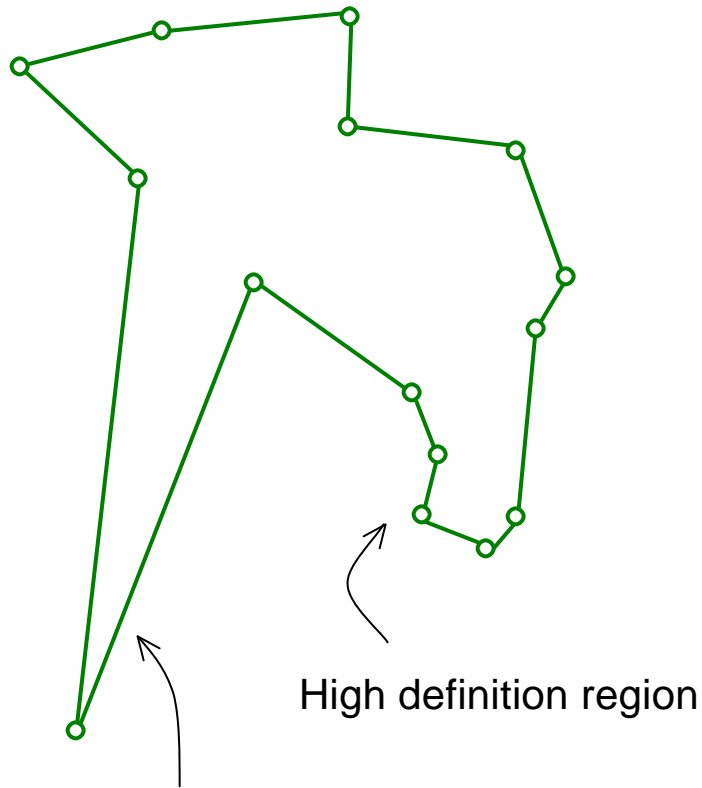
Search this  $2n+m$  dimensional space  
attempting to minimize over the merit function.

The chosen search technique is the simplex method  
of Torczon and Dennis.

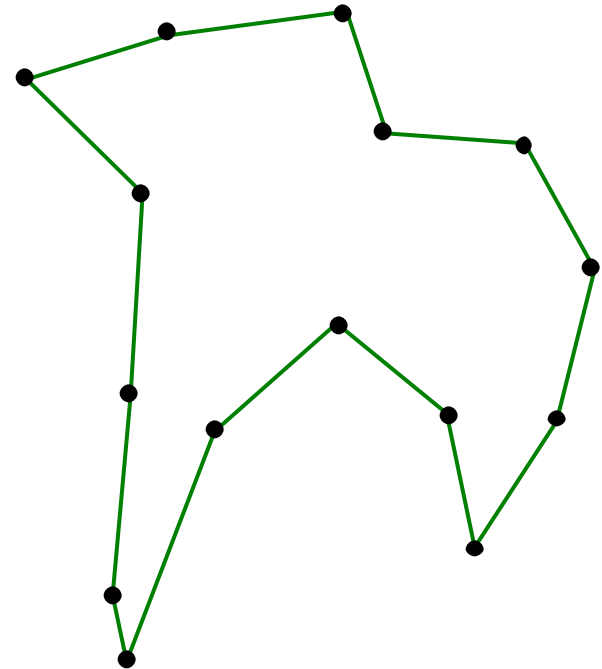


## Example of Object Redefinition (exiting local minima of the merit function)

hypothetical object  
trapped in a local minimum  
of the merit function

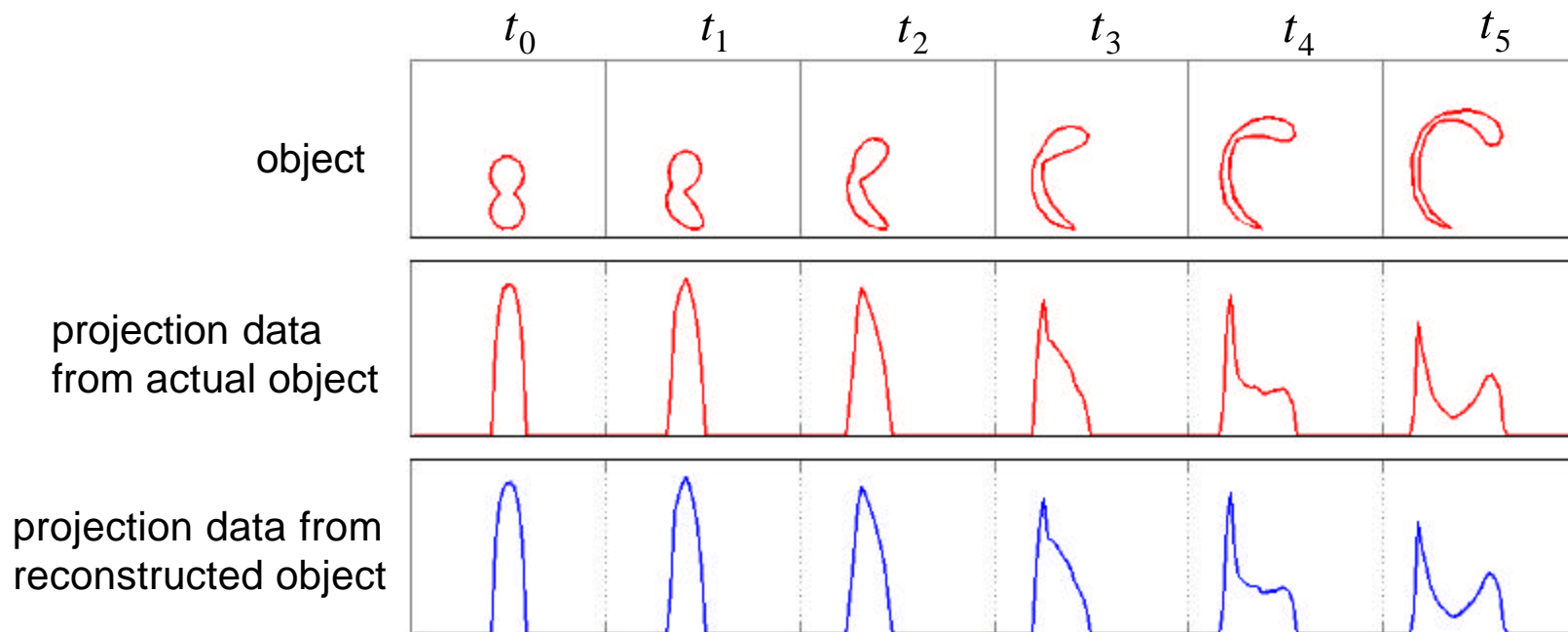


redefined object  
with uniform definition

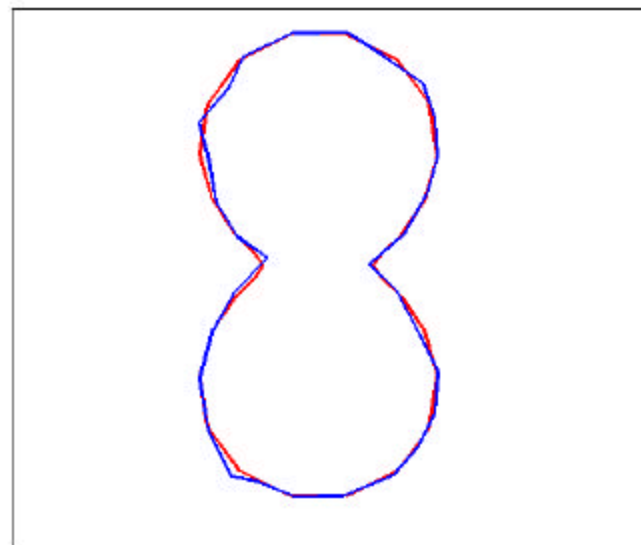


Low-definition region

The merit function will be sensitive to small coordinate changes  
and fine object details will be invisible.

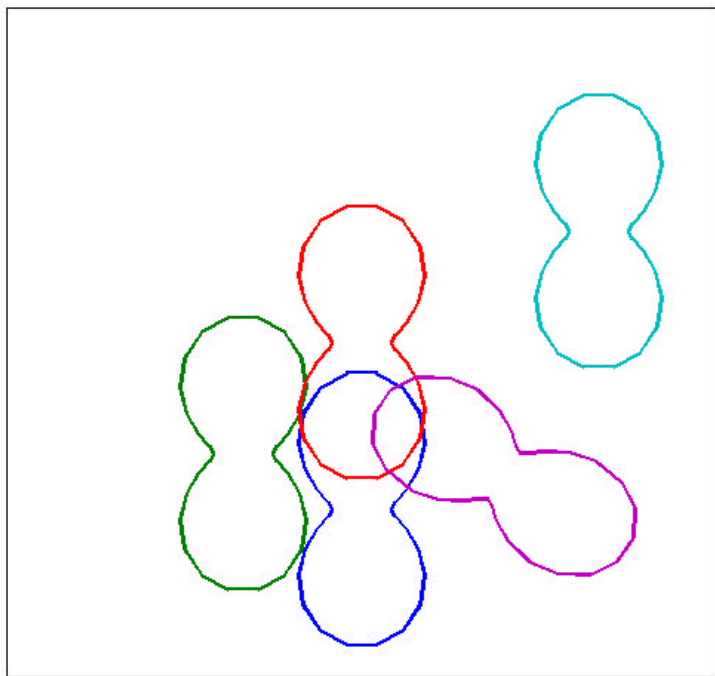


Actual object (red)  
Reconstructed object (blue)

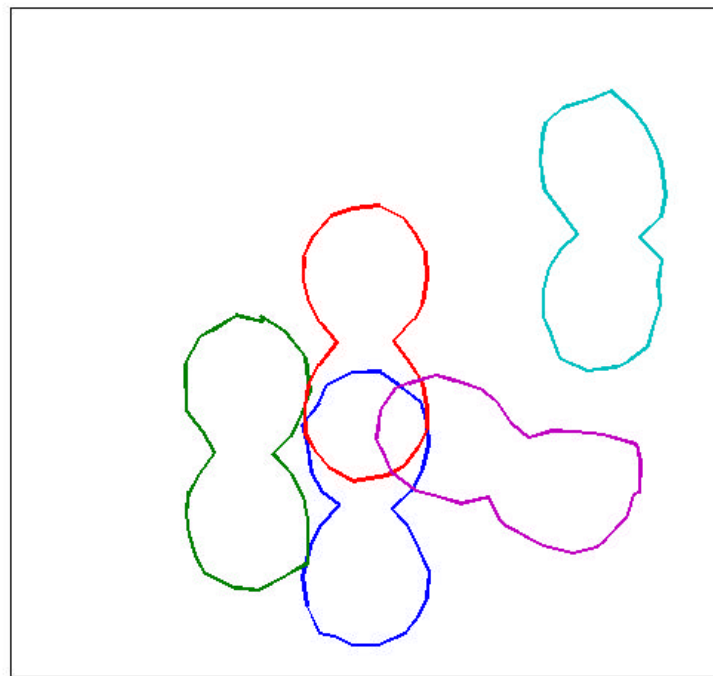


30 point object  
50 sample data sets  
6 time views  
0 parameters in  $F$

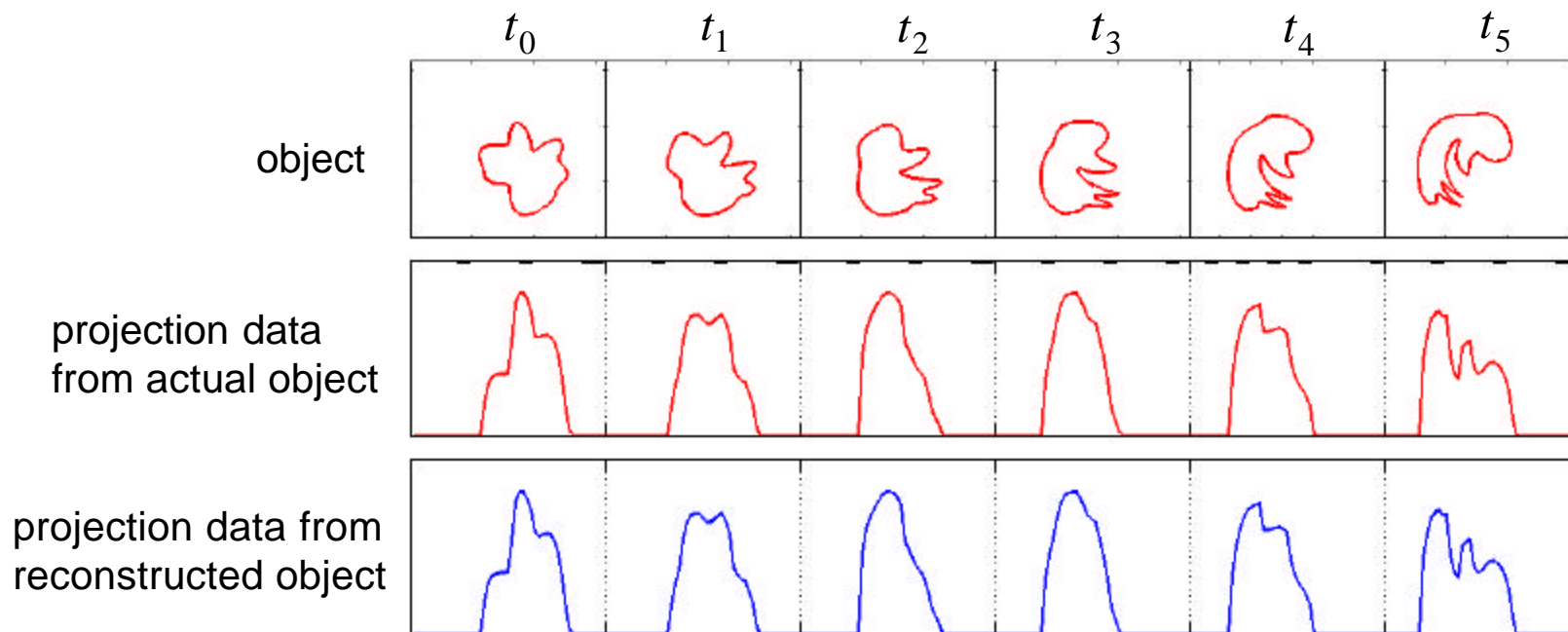
## Identical Objects With Varying Initial Position and Orientation



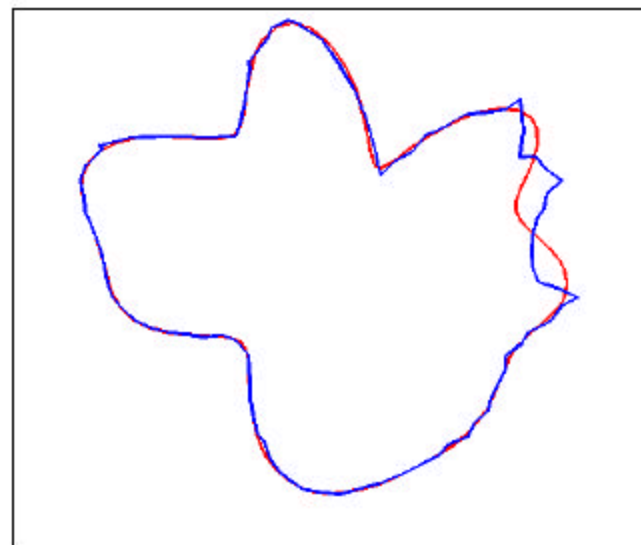
Objects



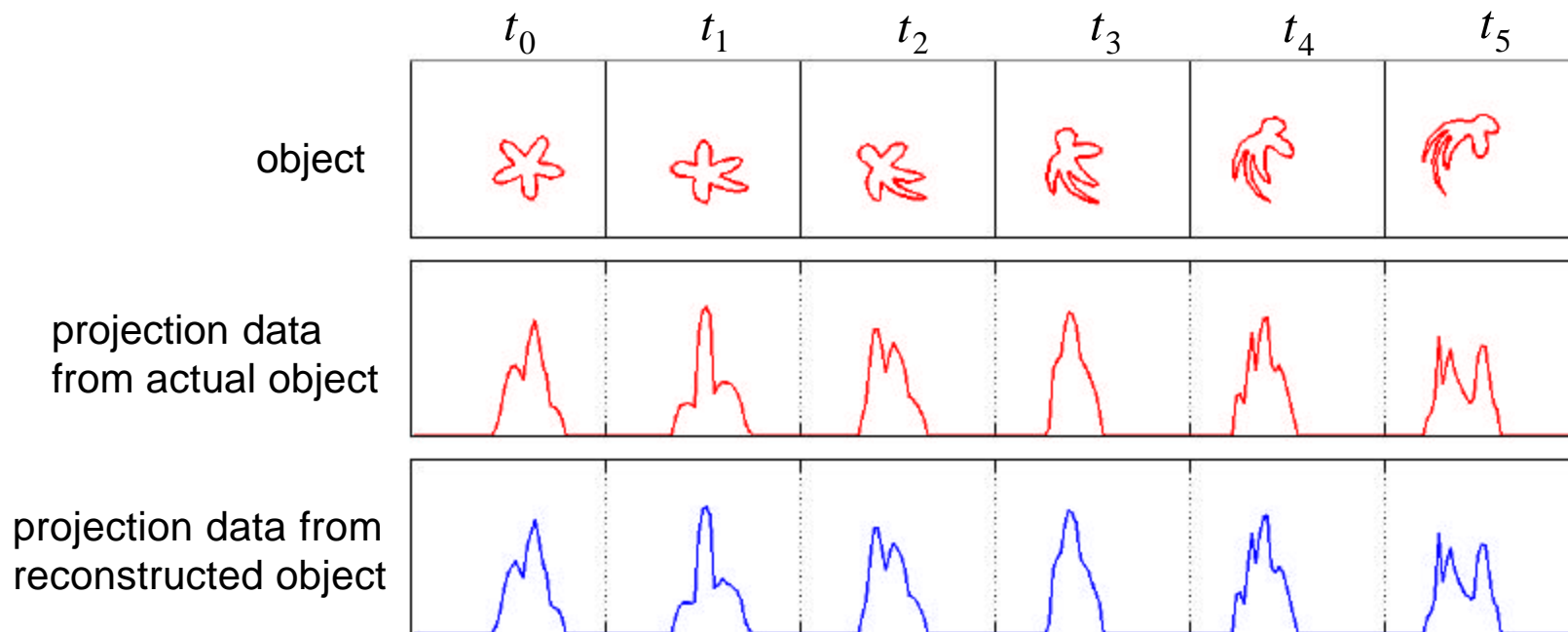
Reconstructions



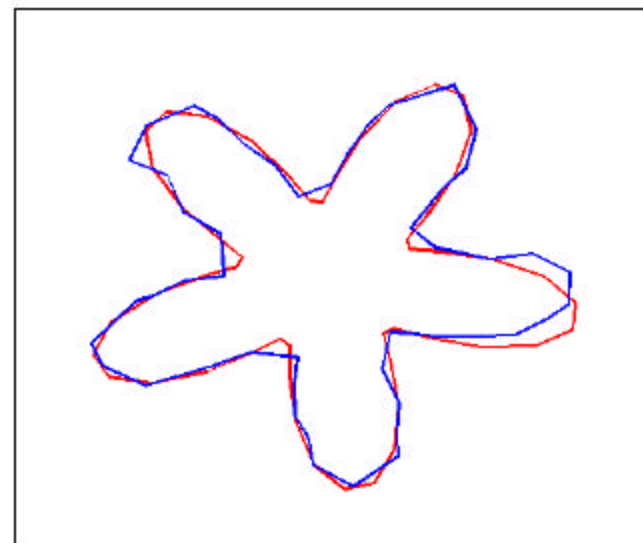
Actual object (red)  
Reconstructed object (blue)



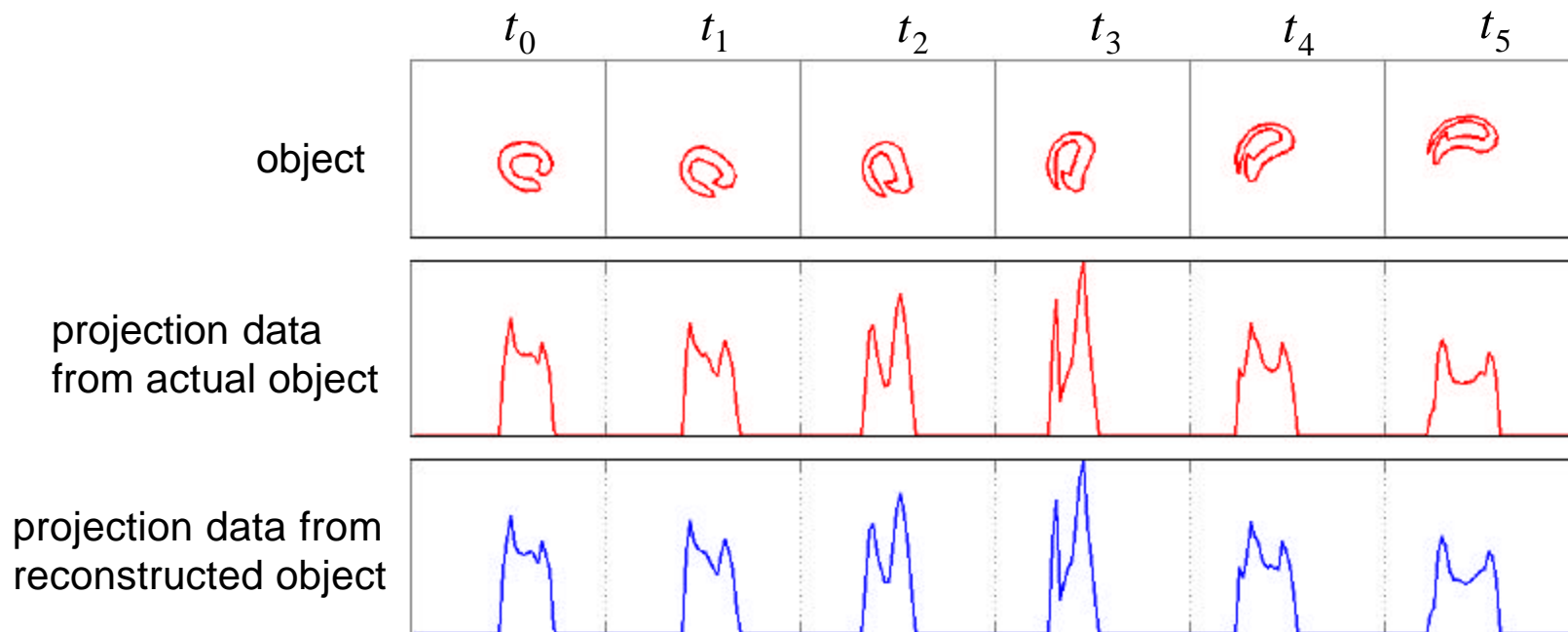
100 point object  
50 sample data sets  
6 time views  
0 parameters in  $F$



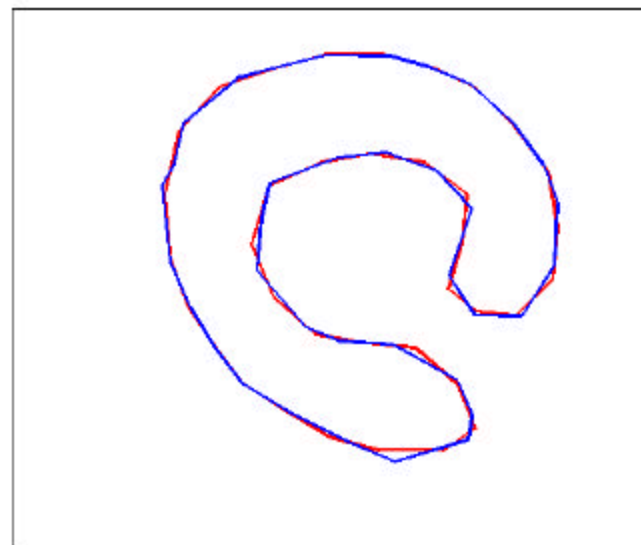
Actual object (red)  
Reconstructed object (blue)



50 point object  
50 sample data sets  
6 time views  
0 parameters in  $F$



Actual object (red)  
Reconstructed object (blue)



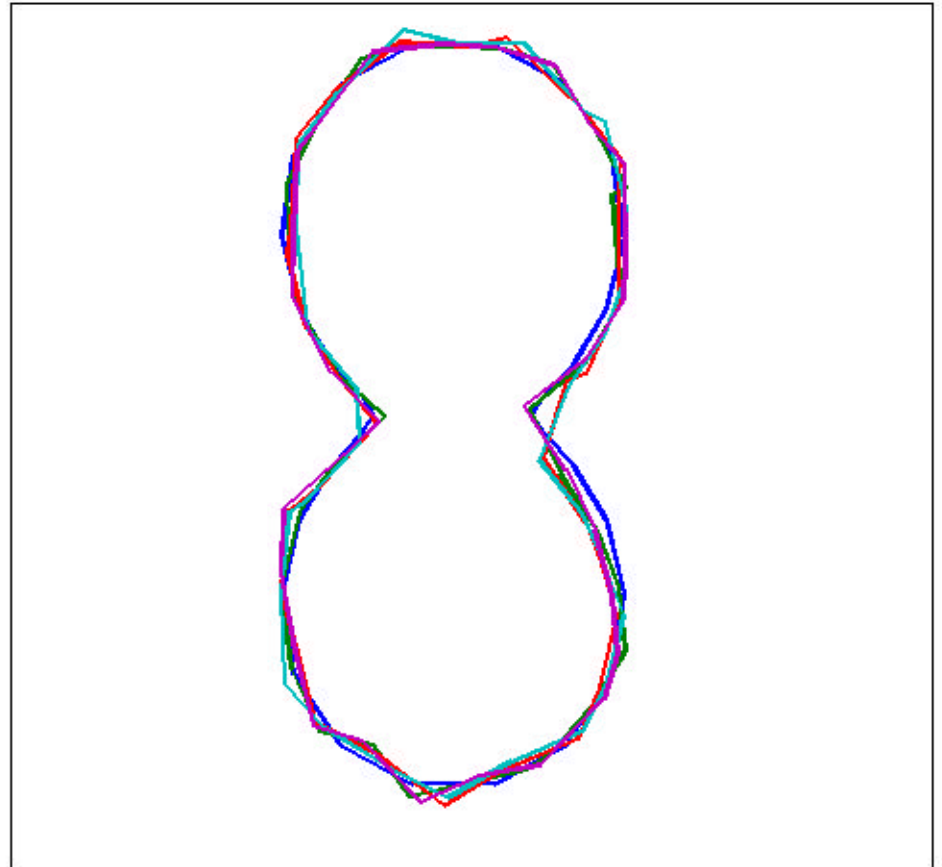
36	point object
50	sample data sets
6	time views
0	parameters in $F$

# Reconstructions From Noisy Data

Simulated Random Particle-Counting Noise  
(e.g. radiographic data)

'Thick' projections have larger noise  
than 'thin' projections.

**Object = blue**  
**1%-5% = green**  
**1%-10% = red**  
**1%-15% = cyan**  
**1%-20% = magenta**

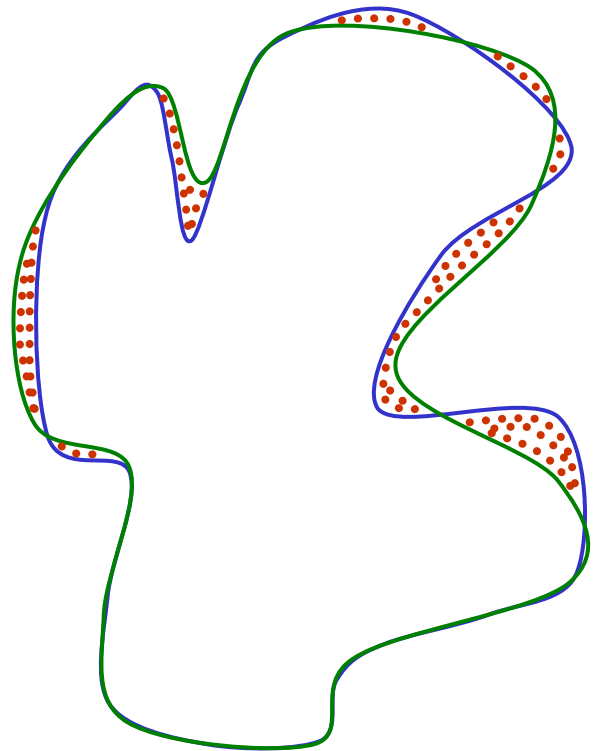


# Creating a Object-Reconstruction Merit Function

Consider an initial object (green) and the best found reconstruction (blue) from dynamics and projection data.

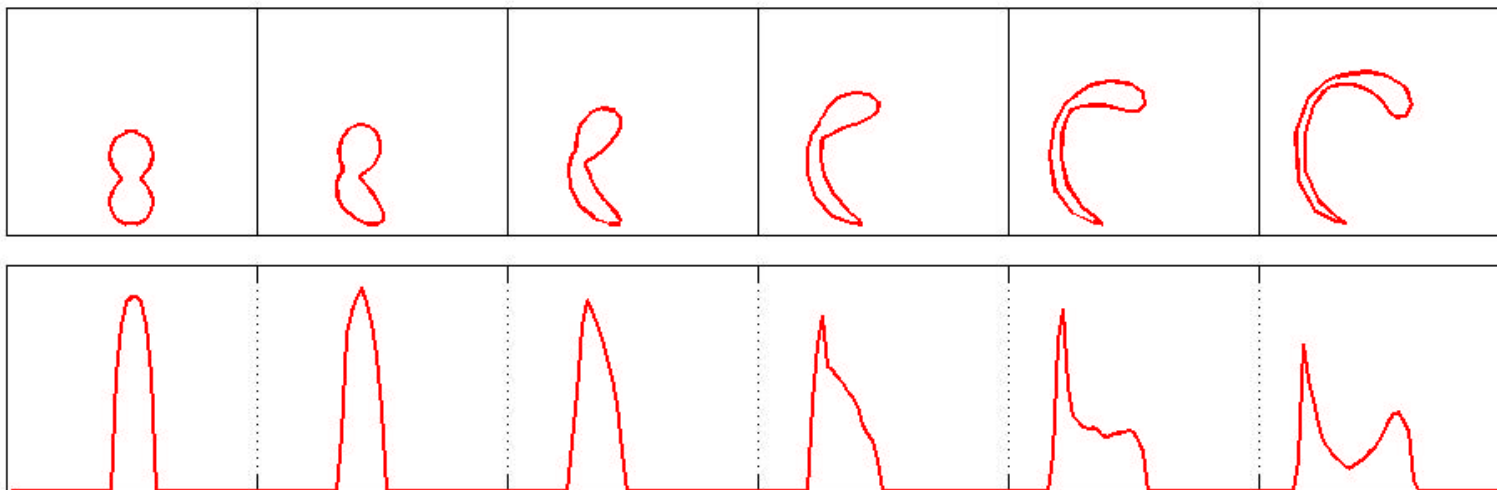
Let the merit function be:

$$g = \frac{\text{difference area (red)}}{\text{object area}}$$





## Reconstruction Quality vs. Number of Views



object points = 30  
data samples per projection = 20  
restart iterations = 30

$$g = \frac{\text{difference area}}{\text{object area}}$$

$$f = \sum_k \|d_k - PF^k z_0\|$$

v	g	f
2	0.372	0.0001
3	0.377	0.0045
4	0.315	0.0078
5	0.245	0.0045
6	0.202	0.0067
7	0.161	0.0050

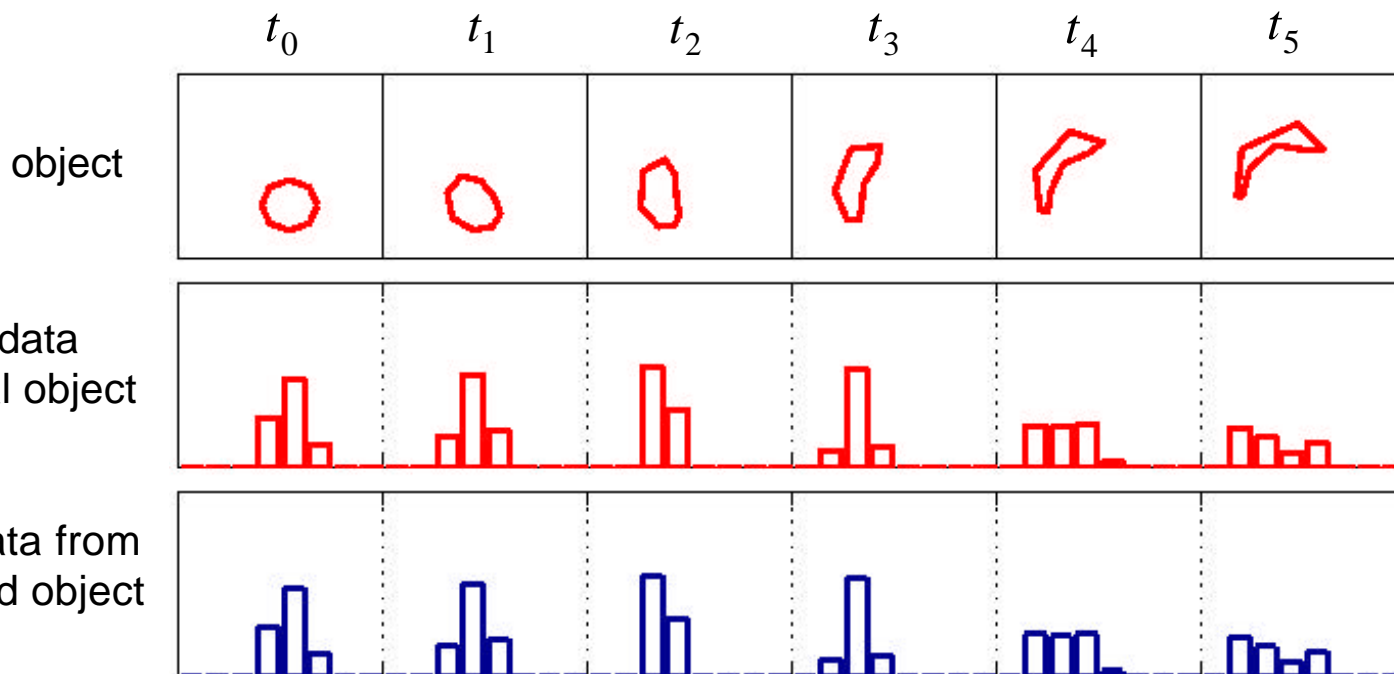
# Estimating Parameters of the Dynamics

Can information about the underlying physics be obtained simultaneously with the object description?

Consider a time evolution operator  $F = F(\lambda, t)$  that is unknown modulo a set of parameters  $\lambda$ . We might modify our advection field ...

$$\dot{x} = -\mathbf{l}_1 \sin^2(x) \sin(2y) + \mathbf{l}_3 + \mathbf{l}_5 x + \mathbf{l}_7 y$$

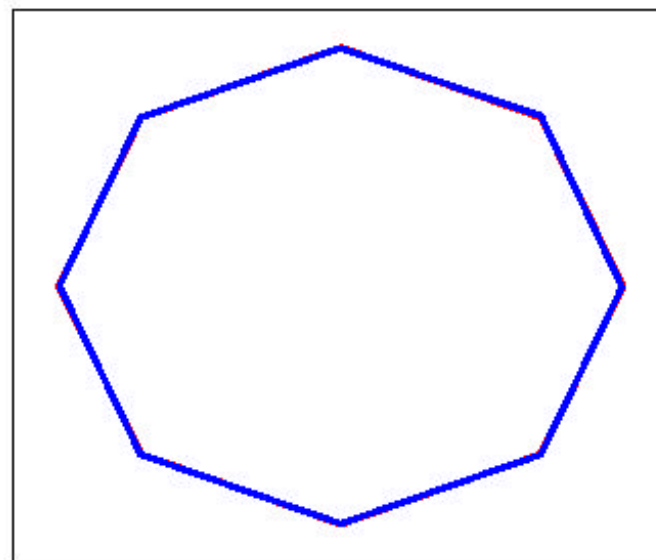
$$\dot{y} = +\mathbf{l}_2 \sin^2(y) \sin(2x) + \mathbf{l}_4 + \mathbf{l}_6 x + \mathbf{l}_8 y$$



8 point object  
 8 sample data sets  
 6 time views  
 8 parameters in  $F$

Actual object (red)  
 Reconstructed object (blue)

$l_1 = +1.000$	$l_2 = +0.999$
$l_3 = +0.000$	$l_4 = +0.000$
$l_5 = -0.000$	$l_6 = +0.000$
$l_7 = -0.000$	$l_8 = -0.000$



object

$t_0$

$t_1$

$t_2$

$t_3$

$t_4$

$t_5$

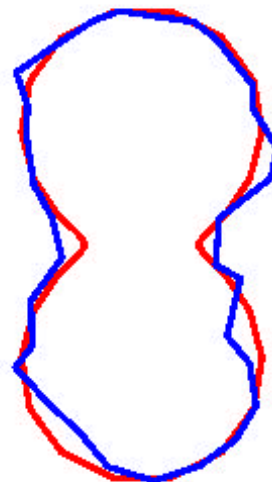
projection data  
from actual object

projection data from  
reconstructed object

30 point object  
30 sample data sets  
6 time views  
8 parameters in  $\mathbf{F}$

Actual object (red)  
Reconstructed object (blue)

$l_1 = +0.982$	$l_2 = +1.019$
$l_3 = -0.001$	$l_4 = +0.010$
$l_5 = +0.005$	$l_6 = -0.003$
$l_7 = -0.006$	$l_8 = -0.003$



## Some Final Thoughts and Directions

Thus far the objects and dynamics are simply defined.  
But the success in recovering them has been very good.

Other object parameterizations have not been explored.

This approach cannot easily handle topological changes in objects.  
One possible path is to explore level-set methods.

Things of some importance not yet done ...

- application to real data
- study of error propagation
- coding optimization
- application to hydrocodes